

Predictive Models for Integrated Manufacturing and Structural Performance of Carbon Fiber Composites for Automotive Applications

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General Motors
2018 Annual Merit Review
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Project ID: MAT117

This presentation does not contain any proprietary, confidential, or otherwise restricted information

Overview



Timeline

Project Start Date: May 1, 2015

Project End Date: April 30, 2019

Percent Complete: 75 %

Budget

Total project funding

• DOE Share: \$6,000,00

• Contractor Share: \$2,571,253

Funding for FY 2017:

DOE Share: \$1,177,715
 Contractor Share: \$504,735

Funding for FY 2018:

DOE share: \$1,820,135

Contractor share: \$780,058

Barriers

- A. Manufacturing Technology: Stochastic manufacturing simulation tools to predict the outcome within 15% of experimental results to reduce cost.
- **B.** Performance Technology: Stochastic structural performance simulation to predict the outcome within 15% of experimental results to optimize design.
- C. Integrated Technology: Integrative manufacturing and structural performance simulation tool that can be used in upfront design to deliver the required assembly performance without any trial and error.

Participants

General Motors
Continental Structural Plastics (CSP)
ESI Group, NA
Altair
University of Southern California

Relevance



Predictive Integrated Modeling Tools

- Primary deliverable: An ICME model capable of predicting stochastic manufacturing and structural performance of carbon fiber (CF) composites.
 - Reduce the cost of manufacturing of CF reinforced automotive components by eliminating trial and error through improved manufacturing simulations.
 - Design, optimize and validate the CF automotive structures in a virtual design space through improved performance modeling.
 - Reduce the lead time and cost to design and implement large scale structural automotive composites.
 - Enable the usage of CF composites for significant light-weighting of automobiles and thus improve fuel economy, and lower emissions, which will reduce greenhouse gas emissions.

Cost Barrier

Will demonstrate the ability to manufacture the automotive CF composites at no more than
 \$4.32 cost per pound weight saved to address the DOE 2030 targets.

Performance Barrier

 Will demonstrate the viability of CF composites to meet vehicle performance requirements while reducing vehicle assembly weight (35% lighter) compared to a current steel design.

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Relevance

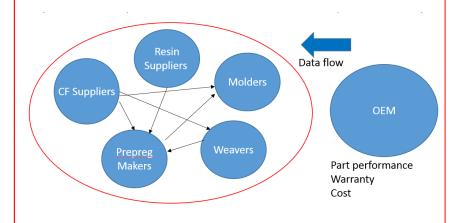
Steps in implementing CF in automobiles Current

Work flow between OEM and Suppliers

Current



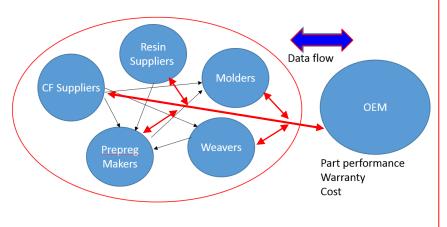
- · Design.
- Selection of manufacturing process.
- Manufacturing feasibility.
- Prototype build and learn.
- Modify design and manufacturing process, if needed.
- Improve prototype build and make part.
- Extrapolate to high volume manufacturing.
- Build the part, iterate to get good quality.
- Evaluate the performance and compare with requirements.
- If failure occurs, redesign the part.



Future

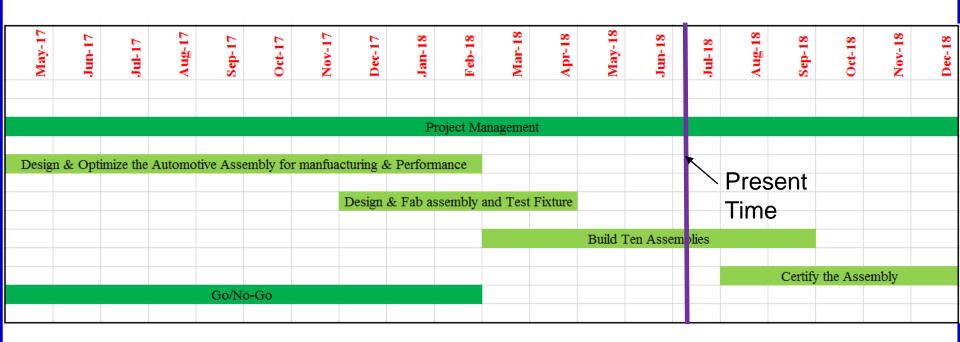
- Design.
- Virtual manufacturing simulation and improve the design for optimizing the cost.
- Include manufacturing outcome in performance simulation and further optimize the design to meet the requirements.
- Build tools, manufacture parts and check the performance

Future



Milestones





All milestones for year 2018 are complete. Go/No-Go decision was also complete.

Approach/Strategy

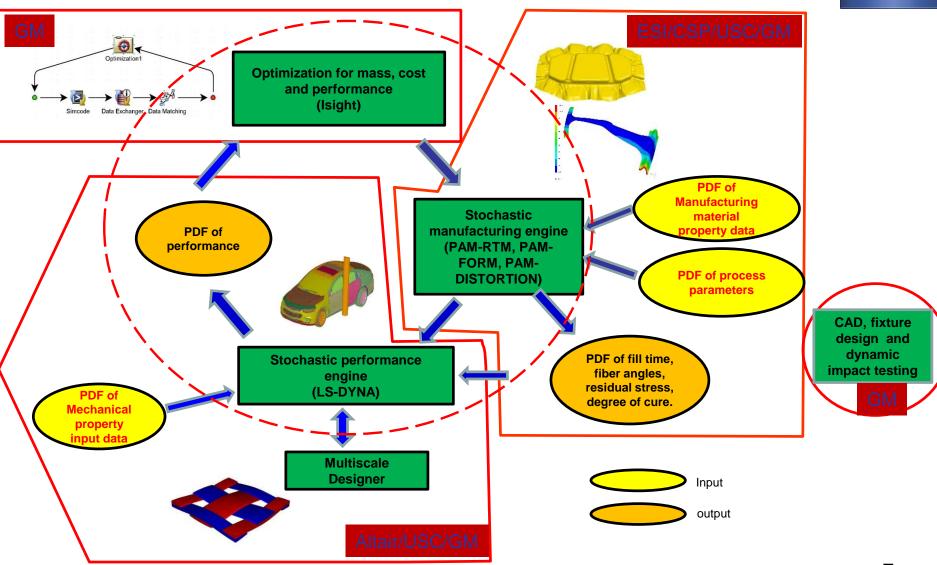


- An ICME approach to develop
 - computational methodologies and tools for predicting stochastic manufacturing.
 - computational methodologies and tools for predicting stochastic performance.
 - Integrated tools to predict the performance of an assembly.
- A team comprised of an automobile OEM, a Tier 1 composite system supplier and molder, software simulation companies in the areas of composite manufacturing and performance prediction, and a DOE funded SciDAC institute for uncertainty quantification.
- Composite System Supplier: Responsible for selecting materials and manufacturing processes for high volume manufacturing, providing plaques and coupons for generating the data required for model calibration and validation.
- Software Companies: Responsible for the development of predictive tools for manufacturing and structural performance
- Stochastic Modeling Research Group: Responsible for developing stochastic models for both manufacturing and structural performance
- OEM: Responsible for developing and conducting experiments for model confirmation, integrating the manufacturing and structural performance tools, demonstrating the technology by design, optimizing, building and testing a carbon fiber automotive assembly as well as validating the developed models by comparing the predictions with experimental results.

Approach/Strategy

GM

Developed a process flow of tool development



Accomplishments



FY 17 Accomplishments

Manufacturing simulation tool development and validation

- Draping model development and validation for non-crimp fabrics.
- Development and validation of resin curing model for state of the art resin from Hexion
- Engineer the HP-RTM process design for two major components and C-RTM process design for other two major components of the automotive assembly planned for demonstration.

Stochastic manufacturing simulation tool development

- Stochastic model results for the complete steps of resin transfer molding draping, injection, curing, etc.
- Complete the development of stochastic manufacturing suite and implement on GM-HPC platform

Structural simulation tool development and validation

- Component validation for a brittle and ductile lay-up
- ICME simulation of truncated pyramid (manufacturing and structural behavior)
- Engineer the structural design for the automotive assembly chosen for the demonstration.

Stochastic structural simulation tool development

- Stochastic structural performance at the component level
- Stochastic structural performance of the automotive assembly

Cost models for the automotive assembly chosen for demonstration

Accomplishments



5 patents submitted to Government Patent office.

Facilities:

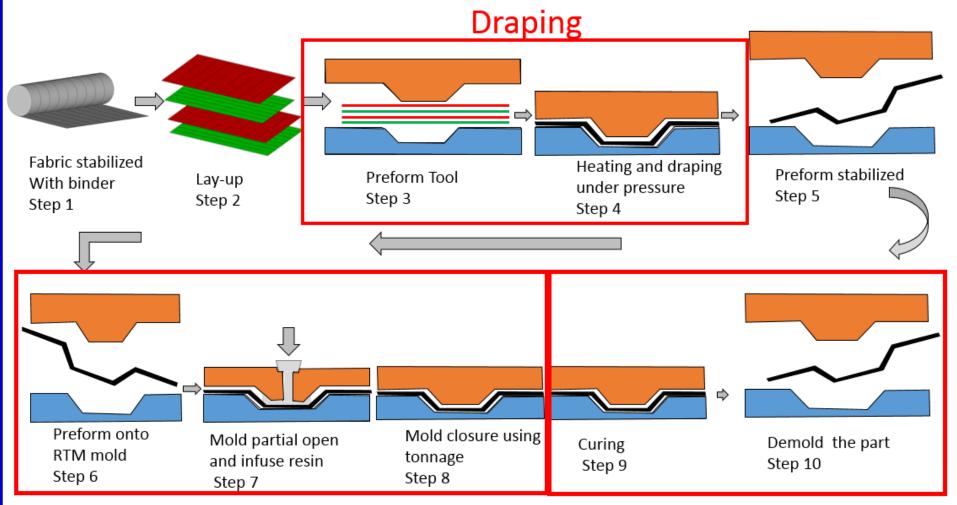
- HP-RTM facility at CSP was upgraded to manufacture components for the assembly.
 The facility is being moved from CSP France to CSP HQ, MI. Three components of the assembly will be molded at the CSP facility.
- HP-RTM facility being installed at GM R&D. One assembly component will be molded at GM.

Outreach:

• To meet the project objectives, strategic agreements were put in place to take the advantage of the state-of-the-art resin from Hexion and novel fabric architectures from Teijin and Chomarat.

Manufacturing Process



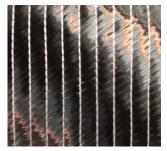


Injection

Curing and Distortion

Draping Simulations

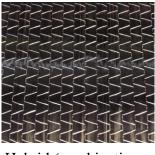
- GM
- Non-crimp fabrics (NCF) offer economic advantages compared to woven fabrics, but are limited by a perceived lack of drapeability.
- Accurate predictability of draping is essential to modify the design variables, processing conditions, and fabric production.
- Stitch pattern Three basic patterns used to make NCFs



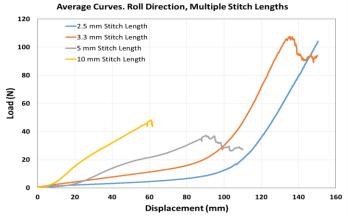
Chain or pillar



Tricot



Hybrid (combination of chain and tricot)



Characterization



2.5 mm Roll



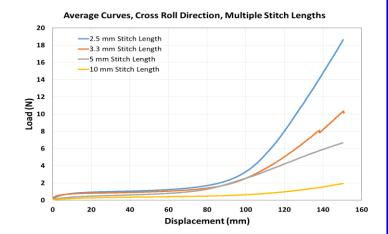
2.5 mm Cross



10 mm Roll

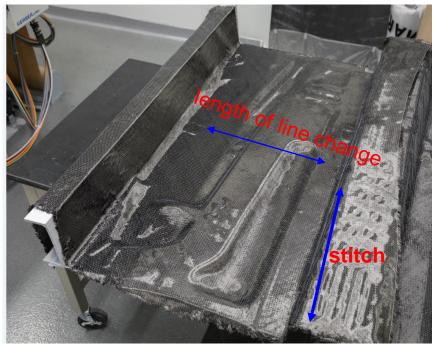


10 mm Cross

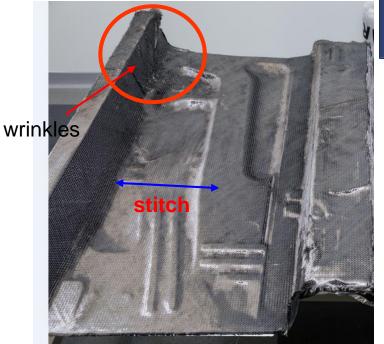


Draping Experiments – Effect of Stich Direction

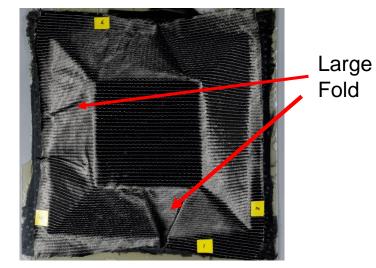




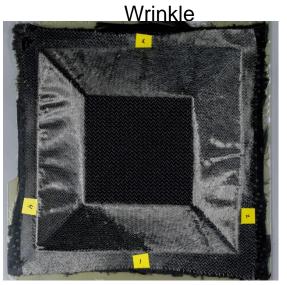
No wrinkles



es Wrir



No Blank Holder



With Blank Holder

240 gsm Veil binder

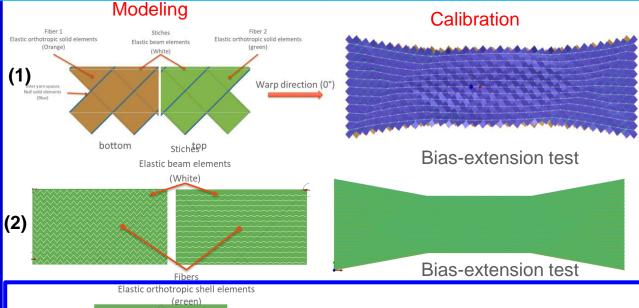
Three Modeling Approaches

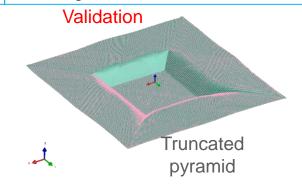


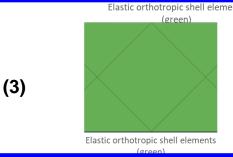
- 1. Meso-scale approach
- 3D solid elements for yarns
- 1D beam elements for stitching
- Contact definition for interaction between yarn and stitching
- 2. Hybrid
- 2D Shell elements for the ply
- 1D beam elements for stitching
- 1D and 2D meshes linked with tied elements

3. Macro-scale

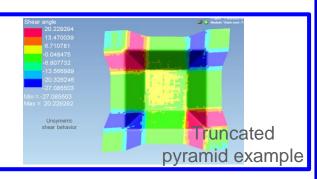
- 2D Shell elements for the ply
- Influence of stitching taken into accounts with different shear behavior for positive and negative shear







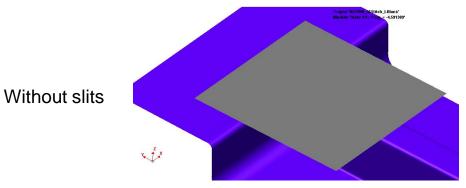


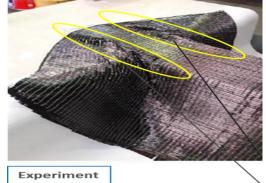


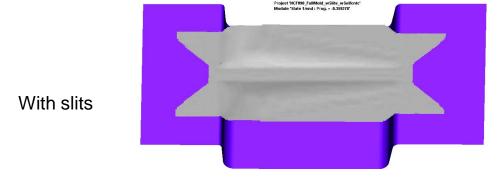
Validation of Draping Models









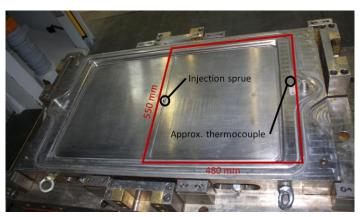




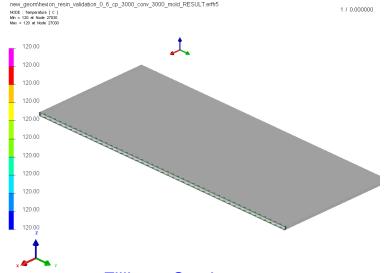
Experiment

Hexion Fast Curing Resin- Model Validation. September 1980 Company of the Company

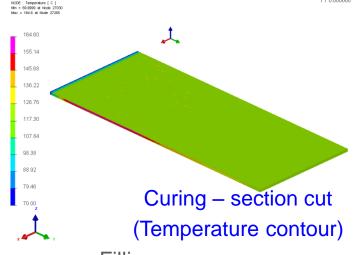


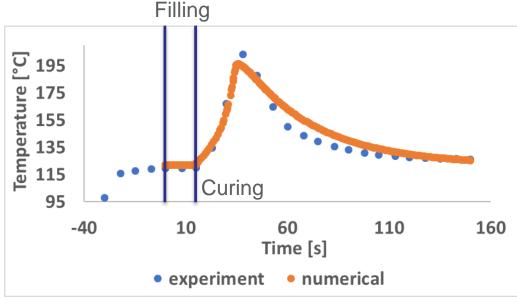


CSP- Europe Experimental setup



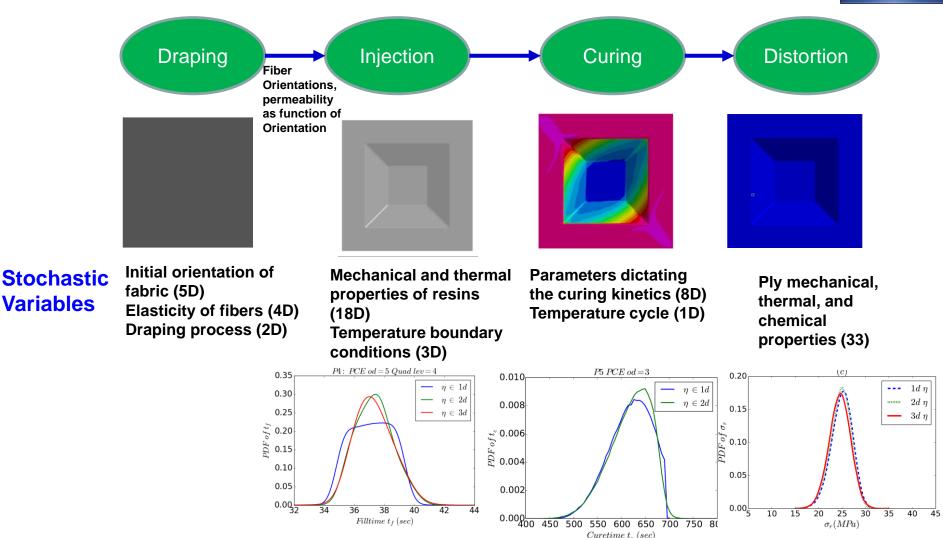
Filling – Section cut (Temperature contour)





Stochastic Manufacturing - Draping, Injection, Curing and Distortion





A total of 74 variables were considered in this integrated problem. This problem is computationally challenging and was solved using state of the art methods developed in this project at USC.

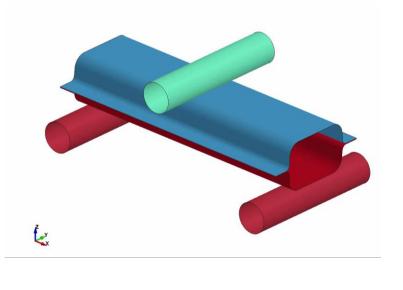
Variables

Structural Modeling – Accomplishments

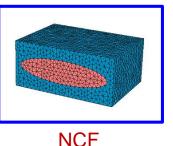
GM

Component validation for a brittle and ductile lay-up ICME simulation of truncated pyramid (manufacturing and structural behavior) Created an engineered structural design for the automotive assembly chosen for the demonstration.

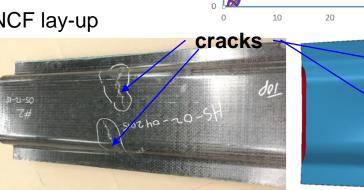
Multi-scale Framework Structural Prediction



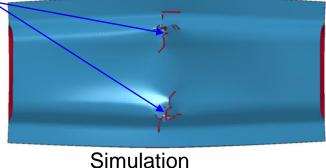
(0/45/-45/90/90/-45/45/0) NCF lay-up



Multi-scale framework



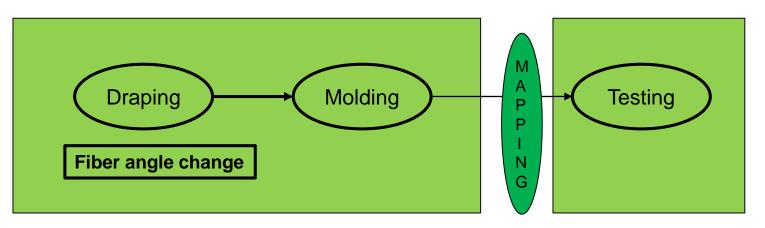
Experiment



Displacement, mm

ICME - Truncated Pyramid





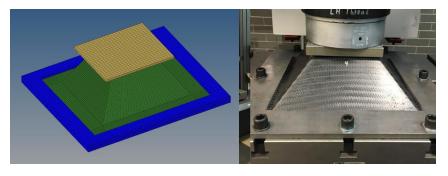


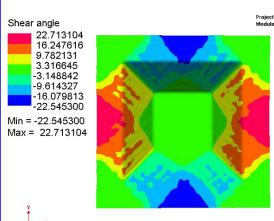


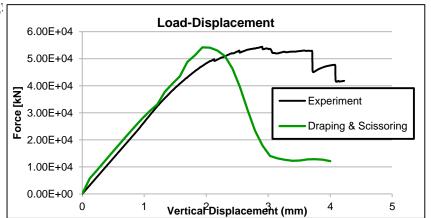
Molding



Structural Performance



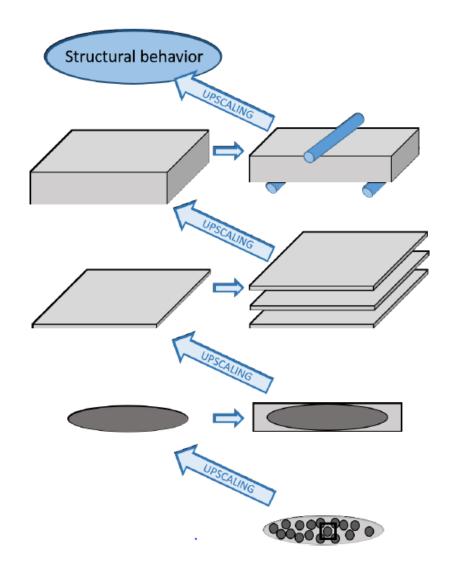




Very good correlations with experiments following ICME process

Stochastic Structural Simulation

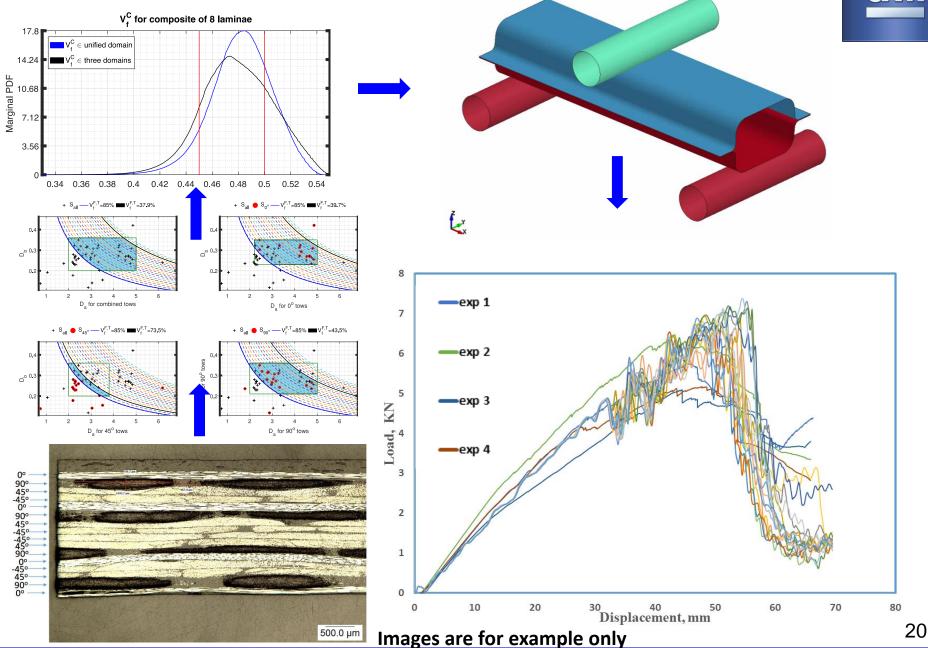




Framework

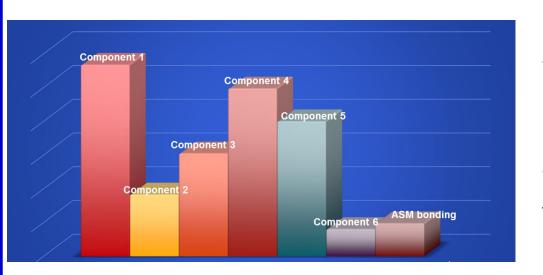
Stochastic Structural Simulation





Cost Modeling





Volume: 80,000/year Product lifetime: 7 years

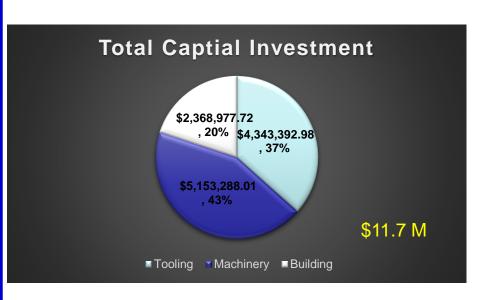
Labor: 3 shifts and 5.4 hours in downtime/day

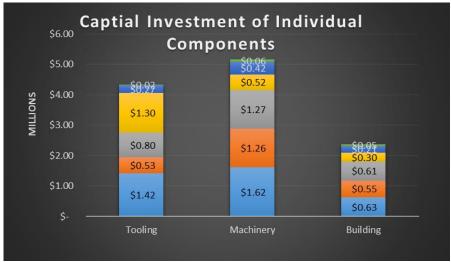
Resin transfer process: HP-RTM ~ 2.8 minutes total cycle time

Capital recovery rate/period: 15%/10 years.

Installation/maintenance:15%/8%

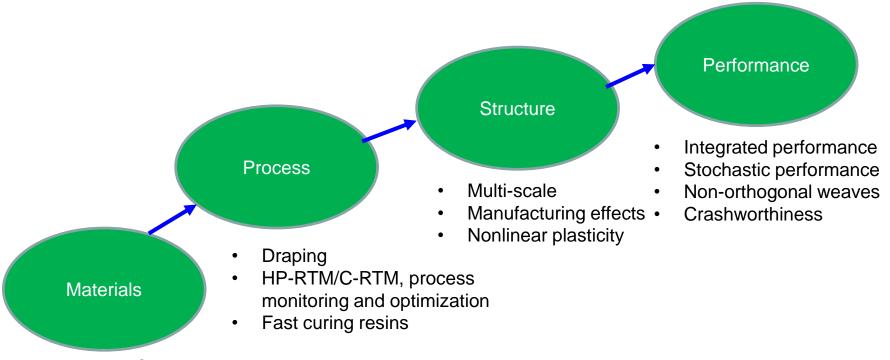
Tool life/materials:1 millions parts, steel base tooling





Current Project - Technology Impact Areas





- Low cost NCF
- Tow size effects (different K)
- Usage of long fiber thermoplastic for energy absorption
- Stochastics at the micro-scale

Responses to Previous Year Reviewers' Comments



1. The reviewer wondered whether a model with stochastic simulations always give the same answer, e.g., for energy absorbed during crash, or instead will a model based on stochastic behavior provide a probability distribution of values as the answer.

Answer: The stochastic simulations will give the performance as probability density function with a mean and variation. This will allow us to understand the variables influencing the variation, tail ends of the distribution so that appropriate actions can be taken.

2. The reviewer replied that a number of presentations have been made and that it would be beneficial to the community if the material models generated in this project could be shared and adapted into various commercial software packages..

Answer: Sure, we are putting together all these developments in the commercial programs so that entire industry can benefit from this project.

Partners/Collaborators



General	Motors	-	Prime

Overall project management, execution, baseline performance evaluation, material data generation for manufacturing and structural simulations, assembly of the CF automotive assembly, testing and validation. material database creation for manufacturing and structural simulation, integrate the manufacturing and structural models, develop cost models, demonstrate the technology development.

Continental Structural Plastics (CSP)

Technology supplier, molder - coupons, plaques and components, develop design for manufacturing guidelines, input for cost models.

ESI Group, NA

Manufacturing simulation models for the manufacturing processes chosen in the project.

Altair

Multi-scale simulation models for the structural performance in the LS-DYNA, ABAQUS and Radioss framework.

University of Southern California

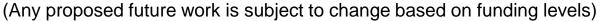
Develop stochastic drivers that work for manufacturing and structural performance simulations. Able to utilize the previous work done on a DOE supported work on uncertainty quantification (SciDAC institute).

Remaining Challenges and Barriers



- Comparison of manufacturing process predictions for the HP-RTM and C-RTM.
- Comparison of structural predictions and experimental results for the crash performance of the assembly built for demonstration.
- Certification of the assembly based on the ICME tools developed in the project.

Proposed Future Research





FY 2018

- Build the tooling required to manufacture the automotive assembly
- Fabricate components and assemble them to test under crash sled.
- Collect the experimental data for the manufacturing (HP-RTM/C-RTM) an structural performance (crush load, damage), etc.stochastically.

FY 2019

- Validation of ICME tool Comparison of prediction and experimental results for manufacturing and structural performance.
- Certification process development.

Summary



- ICME tools stochastic manufacturing and structural performance tool development is complete and implemented successfully on GM-HPC system.
- Several ICME problems were solved and new results show potential for optimizing the process conditions and performance of the composites.
- A large automotive assembly was designed in a virtual space and released for fabrication. Four major components were designed for high volume manufacturing process (HP-RTM and C-RTM).
- Cost models were developed to understand the future potential research areas for economic improvement



Thank You!



Technical Back-Up Slides

Governing Equations in Injection, Curing and Warpage



Filling - Stage - Coupled flow, heat and cure

Darcy's equation – Fluid Flow
$$\nabla \cdot (-\frac{K}{\mu} \overrightarrow{\nabla P}) = 0$$

Heat Transfer Equation
$$\rho C_p \frac{\partial T}{\partial t} + \rho_r C_{pr} V \cdot \nabla T = \nabla \cdot (k \cdot \nabla T) - \rho_r \, \Delta h \, \frac{d\alpha}{dt}$$

Curing Kinetics
$$\frac{d\alpha}{dt} = \left(A_1 \exp\left(-\frac{E_1}{T}\right) + A_2 \exp\left(-\frac{E_2}{T}\right) \cdot \alpha^m\right) \cdot \left(B - \alpha\right)^n$$

Curing - Stage - Coupled heat and cure

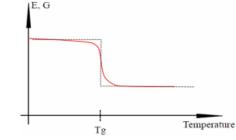
Heat Transfer Equation
$$\rho C_p \frac{\partial T}{\partial t} + \rho_r C_{pr} V \cdot \nabla T = \nabla \cdot (k \cdot \nabla T) - \rho_r \Delta h \frac{d\alpha}{dt}$$
 Curing Kinetics
$$\frac{d\alpha}{dt} = (A_1 \exp\left(-\frac{E_1}{T}\right) + A_2 \exp\left(-\frac{E_2}{T}\right) \cdot \alpha^m) \cdot (B - \alpha)^n$$

Distortion- Stage (Thermo- Chemical Mechanical Analysis)

$$\sigma_{ij}(t) = \int_0^t C_{ijkl}(\xi(t) - \xi(\tau)) \frac{\partial \left(\epsilon_{kl} - \epsilon_{kl}^E\right)}{\partial \tau} d\tau \qquad C_{ijkl}(t) = \begin{cases} 0 & , X < X_{gel} \\ C_{ijkl}^{\infty} + \sum_{p=1}^P C_{ijkl}^{p} \cdot \left(e^{-t/\rho_{ijkl}^P}\right), X \ge X_{gel} \end{cases}, \text{no sum on } i, j, k, l = 0$$

Di Benedetto function $\rightarrow T_g$

$$\frac{T_g - T_{g0}}{T_{g\infty} - T_{g0}} = \frac{\lambda X}{1 - (1 - \lambda)X}$$



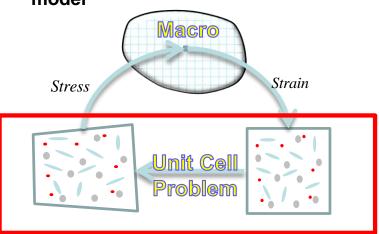
Multiscale Designer Capabilities



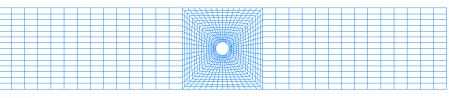
1. Parametric RVE definition

- 1) Geometric scripts
- 2) User-defined parametric RVE
- 3) Integration with experimental data

2. Computational Efficiency: Speed comparable to single scale model



3. Size Effect & Softening after Damage





Challenges:

- (1) Unit cell size comparable to the hole size and much bigger than macro-element size
- (2) Strain softening due to damage

An attempt to account for size effect and softening due to damage

Remedies:

- (1) Rescaling of damage models and
- (2) Staggered nonlocal multiscale approach

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